

# Package: npmla (via r-universe)

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**Type** Package

**Title** Non-Parametric Models for Longitudinal Data Analysis

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**Description** Support the book: Wu CO and Tian X (2018). Nonparametric Models for Longitudinal Data: With Implementation in R. (Chapman & Hall/CRC Monographs on Statistics & Applied Probability); Present global and local smoothing methods for the conditional-mean and conditional-distribution based nonparametric models with longitudinal Data.

**License** GPL (>= 2)

**BugReports** <https://github.com/npmla/npmla/issues>

**URL** <https://github.com/npmla/npmla/>  
<https://www.crcpress.com/Nonparametric-Models-for-Longitudinal-Data-With-Implementation-in-R/Wu-Tian/p/book/9781466516007>

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**Repository** <https://npmla.r-universe.dev>

**RemoteUrl** <https://github.com/npmla/npmla>

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BDIdata	<i>BDIdata dataset</i>
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### Description

This dataset includes 557 depressed patients (total 7117 observations) in the cognitive behavior therapy arm in the Enhancing Recovery in Coronary Heart Disease Patients (ENRICHD) study.

### Usage

```
data(BDIdata)
```

### Format

A data frame with 7117 rows and 5 variables.

**Details**

- ID. Subject ID
- time. Study visit time (in days) since randomization
- BDI. Beck Depression Inventory (BDI) score
- med. Antidepressant medication use
- med.time. The starting time of medication

**References**

1. Wu, C. O., Tian, X. and Bang, H. A varying-coefficient model for the evaluation of time-varying concomitant intervention effects in longitudinal studies. *Statistics in Medicine*, 27:3042-3056, 2008.
2. Wu, C. O., Tian, X. and Jiang, W. A shared parameter model for the estimation of longitudinal concomitant intervention effects. *Biostatistics*, 12(4):737-749, 2011.

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 BMACS

*BMACS CD4 dataset*


---

**Description**

This dataset is from the Baltimore site of the Multi-center AIDS Cohort Study (BMACS), which included 400 homosexual men who were infected by the human immunodeficiency virus (HIV) between 1984 and 1991.

**Usage**

`data(BMACS)`

**Format**

A data frame with 1817 rows and 6 variables.

**Details**

- ID. Subject ID
- Time. Subject's study visit time
- Smoke. Cigarette baseline smoking status
- age. Age at study enrollment
- preCD4. Pre-infection CD4 percentage
- CD4. CD4 percentage at the time of visit

**References**

Kaslow, R. A., Ostrow, D. G., Detels, R., Phair, J. P., Polk, B. F. and Rinaldo, C. R. The Multi-center AIDS Cohort Study: rationale, organization and selected characteristics of the participants. *American Journal of Epidemiology*, 126:310-318, 1987.

---

CVlm *Leave one-subject out Cross-validation score for local linear fit*

---

### Description

Leave one-subject out Cross-validation score for local linear fit

### Usage

CVlm(Xvec, Yvec, bw, ID, Wt)

### Arguments

Xvec, Yvec	numeric vectors of data values, Xvec and Yvec must have the same length.
bw	a bandwidth of the Epanechnikov kernel
ID	subject ID of the data value
Wt	a weight vector, may be subject-specific. a weight vector or a constant. For longitudinal data, $Wt=1/N$ corresponds to measurement uniform weight and $Wt=1/(nni)$ corresponds subject uniform weight.

---

CVspline *Leave one-subject out Cross-validation score for spline fit*

---

### Description

Leave one-subject out Cross-validation score for spline fit

### Usage

CVspline(Xvec, Yvec, ID, nKnots, Degree, Wt)

### Arguments

Xvec, Yvec	numeric vectors of data values, Xvec and Yvec must have the same length.
ID	subject ID of the data value
nKnots	number of equally-spaced knots
Degree	degree of polynomial splines
Wt	a weight vector. For longitudinal data, $Wt=1/N$ corresponds to measurement uniform weight and $Wt=1/(nni)$ corresponds subject uniform weight.

### References

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

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DXi	<i>Derivative of the function Xi(s)</i>
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---

**Description**

Derivative of the function Xi(s)

**Usage**

DXi(s)

**Arguments**

s                    a number or a vector

**Value**

value of the function DXi with give s

**Examples**

```
DXi(c(-1000, -10, -5, 0, 5, 10, 1000 ))
```

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HSCT	<i>HSCT dataset</i>
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**Description**

This dataset consists of 20 patients with hematologic malignancies who had allogeneic hematopoietic stem cell transplantation (HSCT) between 2006 and 2009 at the National Institutes of Health (NIH). The variables are as follows:

**Usage**

```
data(HSCT)
```

**Format**

A data frame with 271 rows and 8 variables.

**Details**

- ID. Subject ID
- Days. Subject's study visit time relative to time of transplant (day 0)
- Granu. Granulocytes (K/uL)
- LYM. Lymphocytes (K/uL)
- MON. Monocytes (K/uL)
- G-CSF. Granulocyte colony-stimulating factor level (pg/mL)
- IL-15. IL-15 level (pg/mL)
- MCP-1. monocyte chemotactic protein-1 level (pg/mL)

**References**

Melenhorst, J.J., Tian, X., Xu, D., Sandler, N.G., Scheinberg, P., Biancotto, A., et al. Cytopenia and leukocyte recovery shape cytokine fluctuations after myeloablative allogeneic hematopoietic stem cell transplantation. *Haematologica*, 97(6):867-73, 2012.

---

kernel.fit

*Nadaraya-Watson Kernel estimator*


---

**Description**

Nadaraya-Watson Kernel estimator

**Usage**

```
kernel.fit(Xint, Xvec, Yvec, bw, Kernel = "Ep", Wt = 1)
```

**Arguments**

Xint	a vector of x interval to generate the local linear fit
Xvec, Yvec	numeric vectors of data values, Xvec and Yvec must have the same length.
bw	a bandwidth of the kernel
Kernel	a character string indicating which kernel function is to be used. Use of "Ep", "Bw", or "Nm" for Epanechnikov, Biweight or Normal kernel function.
Wt	a weight vector

**References**

1. Fan, J. and Gijbels, I. *Local Polynomial Modeling and Its Applications*. Chapman & Hall, London, United Kingdom, 1996.
2. Wu, C.O. and Tian, X. *Nonparametric Models for Longitudinal Data: With Implementation in R*. Chapman & Hall/CRC. 2018

**Examples**

```
X <- seq(0, 1, len=100)
Y <- (X- 0.5)^3 - 2*(X-0.5)^2+ rnorm(100, 0, 0.1)
kernel.fit(seq(0,1,0.1), X, Y, Kernel="Ep", bw=0.1, Wt=1 )
```

---

Kernel2D

*2-dim Kernel function for longitudinal data*


---

**Description**

2-dim Kernel function for longitudinal data

**Usage**

```
Kernel2D(ID1s, Xvec, Yvec, X01, X02, Bndwdth1, Bndwdth2)
```

**Arguments**

ID1s                    the vector of subject ID in a longitudinal sample  
Xvec                    Yvec numeric vectors of data values, Xvec and Yvec must have the same length  
X01                    X02 two given values of Xvec  
Bndwdth1, Bndwdth2    two given bandwidths

**Value**

2-dim kernel fit result

**References**

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

---

Kernel3D

*3-dim Kernel function for longitudinal data to get  
 $Pr(y1(t1),y2(t2)|x(t1))$* 


---

**Description**

3-dim Kernel function for longitudinal data to get  $Pr(y1(t1),y2(t2)|x(t1))$

**Usage**

```
Kernel3D(ID1s = ID, Y, Time, X, T1, T2, X0, Bndwdth1, Bndwdth2, Bndwdth3)
```

**Arguments**

ID1s	the vector of subject ID in a longitudinal sample
Y, X, Time	numeric vectors of outcome, covariate and time of the the same length
T1, T2	twp given time points
X0	a given covariate value
Bndwdth1, Bndwdth2, Bndwdth3	three bandwidths around two time and one covariate value

**Value**

3-dim Kernel function results

**References**

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

---

Kernel3D.S2

*3-dim Kernel function for longitudinal data to get  $Pr(y_2(t_2)|x(t_1))$*

---

**Description**

3-dim Kernel function for longitudinal data to get  $Pr(y_2(t_2)|x(t_1))$

**Usage**

Kernel3D.S2(ID1s = ID, Y, Time, X, T1, T2, X0, Bndwdth1, Bndwdth2, Bndwdth3)

**Arguments**

ID1s	the vector of subject ID in a longitudinal sample
Y, X, Time	numeric vectors of outcome, covariate and time of the the same length
T1, T2	twp given time points
X0	a given covariate value
Bndwdth1, Bndwdth2, Bndwdth3	three bandwidths around two time and one covariate value

**Value**

3-dim Kernel function results

**References**

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.



---

Kh.Bw                      *Biweight kernel*

---

**Description**

Biweight kernel

**Usage**

Kh.Bw(datavec, Bndwdth)

**Arguments**

datavec                  a numeric vector  
Bndwdth                  a bandwidth of the kernel

**Value**

kernel function result

**Examples**

# same usage as Kh.Ep

---

Kh.Ep                      *Epanechnikov Kernel*

---

**Description**

Epanechnikov Kernel

**Usage**

Kh.Ep(datavec, Bndwdth)

**Arguments**

datavec                  a numeric vector  
Bndwdth                  a bandwidth

**Value**

kernel function result

**Examples**

Kh.Ep(2:7,5)

Kh.Nm

*Normal kernel*

---

**Description**

Normal kernel

**Usage**

Kh.Nm(datavec, Bndwdth)

**Arguments**

datavec	a numeric vector
Bndwdth	a bandwidth of the kernel

**Value**

kernel function result

**Examples**Kh.Nm(2:7,5)

---

Kh2D

*Multiplicative Epanechnikov Kernel (2-dim)*

---

**Description**

Multiplicative Epanechnikov Kernel (2-dim)

**Usage**

Kh2D(datavec1, datavec2, Bndwdth1, Bndwdth2)

**Arguments**

datavec1	datavec2 two numeric vectors of same length
Bndwdth1	Bndwdth2 two bandwidths for two vectors

**Value**

2-dim kernel function result

**Examples**

Kh2D(2:7, 2:7, 5, 5)

---

Kh3D *Multiplicative Epanechnikov Kernel (3-dim)*

---

**Description**

Multiplicative Epanechnikov Kernel (3-dim)

**Usage**

```
Kh3D(datavec1, datavec2, datavec3, Bndwdth1, Bndwdth2, Bndwdth3)
```

**Arguments**

datavec1	datavec2, datavec3 three numeric vectors of same length
Bndwdth1	Bndwdth2, Bndwdth3 three bandwidths for three vectors

**Value**

3-dim kernel function result

---

LocalLm *Local linear fit with Epanechnikov kernel*

---

**Description**

Local linear fit with Epanechnikov kernel

**Usage**

```
LocalLm(Xint, Xvec, Yvec, bw, Wt = 1)
```

**Arguments**

Xint	a vector of x interval to generate the local linear fit
Xvec, Yvec	numeric vectors of data values, Xvec and Yvec must have the same length.
bw	a bandwidth of the kernel
Wt	a weight vector

**Examples**

```
data(BMACS)
Time.int<- seq(0.1,5.9, by=0.1)
LocalFit.Y <- with(BMACS, LocalLm(Time.int, Time, CD4, bw=0.9, Wt=1))
```

---

LocalLm.Beta	<i>Least square local linear fit</i>
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---

### Description

Least square local linear fit

### Usage

```
LocalLm.Beta(Tint, Tvec, X1, X2, X3, Yvec, Bndwdth, Weight)
```

### Arguments

Tint	a time interval
Tvec, Yvec	numeric vectors of time and outcome values, Tvec and Yvec must have the same length.
X1, X2, X3	three covariate vectors
Bndwdth	a bandwidth of the Epanechnikov kernel
Weight	the weight vector

### References

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

### Examples

```
data(NGHS)
NGHS$Black <- (NGHS$RACE==2)*1
NGHS<- NGHS[!is.na(NGHS$SBP) & !is.na(NGHS$BMIPCT) & !is.na(NGHS$HTPCT ),]
Ct <- data.frame(table(NGHS$ID))
names(Ct)<- c('ID', 'ni')
NGHS<- merge(NGHS, Ct, by= 'ID')
nID<- dim(Ct)[1]
Age.grid <- seq(9, 19, by=0.5) #21
NGHS$HTPCTc<- NGHS$HTPCT-50
NGHS$BMIPCTc<- NGHS$BMIPCT-50
Beta <- with(NGHS, LocalLm.Beta(Age.grid, AGE, X1=Black, X2=HTPCTc, X3=BMIPCTc, SBP, Bndwdth=3.5, Weight=1/ni))
```

---

LocalLm.Beta.t0      *Least square local linear fit at t0*

---

**Description**

Least square local linear fit at t0

**Usage**

LocalLm.Beta.t0(t0, Tvec, X1, X2, X3, Yvec, Bndwdth, Weight)

**Arguments**

t0	a given time point
Tvec, Yvec	numeric vectors of time and outcome values, Tvec and Yvec must have the same length.
X1, X2, X3	three covariate vectors
Bndwdth	a bandwidth of the Epanechnikov kernel
Weight	the weight vector

**References**

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

**Examples**

```
# see usage of LocalLm.Beta
```

---

LocalLm.X0      *Local linear fit at X0 with Epanechnikov kernel*

---

**Description**

Local linear fit at X0 with Epanechnikov kernel

**Usage**

LocalLm.X0(Xvec, Yvec, X0, Bndwdth, Wt = 1)

**Arguments**

Xvec, Yvec	numeric vectors of data values, Xvec and Yvec must have the same length.
X0	a given value
Bndwdth	a bandwidth of the kernel
Wt	a weight vector or a constant. For longitudinal data, $Wt=1/N$ corresponds to measurement uniform weight and $Wt=1/(n_i)$ corresponds subject uniform weight.

**Examples**

```
# see usage of LocalLm
```

---

Newton1var	<i>An equation solver with Newton's method with 1 variable</i>
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---

**Description**

An equation solver with Newton's method with 1 variable

**Usage**

```
Newton1var(Z12vec, h0, Vh, HZB, Ind.Y, Diff = 1e-08, ORR, MaxIter = 100)
```

**Arguments**

Z12vec	2-dim covariate vector
h0, Vh, HZB	inital values
Ind.Y	outcome inidicator
Diff	limit to stop the interations
ORR	estimate of the odds ratio vector
MaxIter	maximum no. of interations

**Value**

The root of the equation

**References**

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

---

Newton2var	<i>An equation solver with Newton's method with 2 variables</i>
------------	---

---

**Description**

An equation solver with Newton's method with 2 variables

**Usage**

```
Newton2var(Zij, b0, Ub, Indicator, diff1mt = 1e-14, MaxIter = 100)
```

**Arguments**

Zij	2-dim covariate vector
b0, Ub	inital values
Indicator	Indicator of $Y_{i1} > Y_{i2}$
diff1mt	limit to stop the interations
MaxIter	maximum no. of interations

**Value**

The root of the equation

**References**

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

---

NGHS	<i>NGHS dataset</i>
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**Description**

This dataset includes 2378 girls (total 19701 observations) enrolled in the National Heart, Lung, and Blood Institute's Growth and Health Study (NGHS). NGHS is a multicenter population-based cohort study aimed at evaluating the racial differences and longitudinal changes in childhood cardiovascular risk factors between Caucasian and African American girls during childhood and adolescence.

**Usage**

```
data(NGHS)
```

**Format**

A data frame with 19701 rows and 12 variables.

**Details**

- ID. Subject ID
- RACE. Subject's race (1=Caucasian, 2= African American)
- AGE,HEIGHT,WEIGHT,BMI. Age, height, weight and BMI at study visit
- BMIPCT, HTPCT. CDC Age-adjusted BMI percentile and height percentile at study visit
- SBP,DBP. Systolic and diastolic blood pressure at study visit
- TG,LDL. Triglyceride and Low-density lipoprotein (LDL) cholesterol at study visit

**References**

1. National Heart, Lung, and Blood Institute Growth and Health Research Group (NGHSRG). Obesity and cardiovascular disease risk factors in black and white girls: the NHLBI Growth and Health Study. *American Journal of Public Health*, 82:1613-1620, 1992.
2. Wu, C. O. and Tian, X. Nonparametric estimation of conditional distributions and rank-tracking probabilities with time-varying transformation models in longitudinal studies. *Journal of the American Statistical Association*, 108:971-982, 2013.

---

 NW.Kernel

---

*Title Nadaraya-Watson Kernel estimator at x0*


---

**Description**

Title Nadaraya-Watson Kernel estimator at x0

**Usage**

NW.Kernel(Xvec, Yvec, X0, Kernel = "Ep", Bndwidth, Wt = 1)

**Arguments**

Xvec, Yvec	numeric vectors of data values, Xvec and Yvec must have the same length.
X0	a given value
Kernel	a character string indicating which kernel function is to be used. Use of "Ep", "Bw", or "Nm" for Epanechnikov, Biweight or Normal kernel function.
Bndwidth	a bandwidth of the kernel
Wt	a weight vector or a constant. For longitudinal data, Wt=1/N corresponds to measurement uniform weight and Wt=1/(nni) corresponds subject uniform weight.

**Value**

The kernel estimator at x0



## References

1. Fan, J. and Gijbels, I. Local Polynomial Modeling and Its Applications. Chapman & Hall, London, United Kingdom, 1996.
2. Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data. Chapman & Hall/CRC. To appear.

## Examples

```
X <- seq(0, 1, len=100)
Y <- (X- 0.5)^3 - 2*(X-0.5)^2+ rnorm(100, 0, 0.1)
NW.WtKernel(X, Y, X0=0.5, Kernel="Ep", Bndwdth=0.3, Wt=1 )
NW.WtKernel(X, Y, X0=0.5, Kernel="Nm", Bndwdth=0.3, Wt=1 )
```

---

NW.WtKernel

*Title Nadaraya-Watson Kernel estimator at x0*

---

## Description

Title Nadaraya-Watson Kernel estimator at x0

## Usage

```
NW.WtKernel(Xvec, Yvec, X0, Kernel = "Ep", Bndwdth, Wt = 1)
```

## Arguments

Xvec, Yvec	numeric vectors of data values, Xvec and Yvec must have the same length.
X0	a given value
Kernel	a character string indicating which kernel function is to be used. Use of "Ep", "Bw", or "Nm" for Epanechnikov, Biweight or Normal kernel function.
Bndwdth	a bandwidth of the kernel
Wt	a weight vector or a constant. For longitudinal data, Wt=1/N corresponds to measurement uniform weight and Wt=1/(nni) corresponds subject uniform weight.

## Value

The kernel estimator at x0

## References

1. Fan, J. and Gijbels, I. Local Polynomial Modeling and Its Applications. Chapman & Hall, London, United Kingdom, 1996.
2. Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018

**Examples**

```
X <- seq(0, 1, len=100)
Y <- (X- 0.5)^3 - 2*(X-0.5)^2+ rnorm(100, 0, 0.1)
NW.WtKernel(X, Y, X0=0.5, Kernel="Ep", Bndwdth=0.3, Wt=1 )
NW.WtKernel(X, Y, X0=0.5, Kernel="Nm", Bndwdth=0.3, Wt=1 )
```

---

<code>spline.fit</code>	<i>Polynomial-spline fit with equally-spaced knots</i>
-------------------------	--

---

**Description**

Polynomial-spline fit with equally-spaced knots

**Usage**

```
spline.fit(Xint, Xvec, Yvec, nKnots = 2, Degree = 3, Wt = 1)
```

**Arguments**

<code>Xint</code>	a vector of x interval to generate the local linear fit
<code>Xvec, Yvec</code>	numeric vectors of data values, Xvec and Yvec must have the same length.
<code>nKnots</code>	number of equally-spaced knots
<code>Degree</code>	degree of polynomial splines
<code>Wt</code>	a weight vector or a constant. For longitudinal data, $Wt=1/N$ corresponds to measurement uniform weight and $Wt=1/(nni)$ corresponds subject uniform weight.

**References**

Wu, C.O. and Tian, X. Nonparametric Models for Longitudinal Data: With Implementation in R. Chapman & Hall/CRC. 2018.

---

<code>Xi</code>	<i>Function Xi(s)</i>
-----------------	-----------------------

---

**Description**

Function Xi(s)

**Usage**

```
Xi(s)
```

**Arguments**

<code>s</code>	a number or a vector
----------------	----------------------

**Value**

value of the function with give s

**Examples**

$X_i(0)$

$X_i(c(-1000, -10, -5, 0, 5, 10, 1000 ))$

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